

# Using expert's knowledge in Bayesian analysis

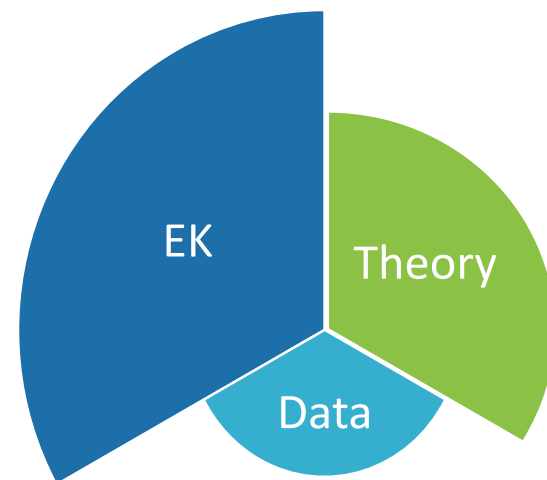
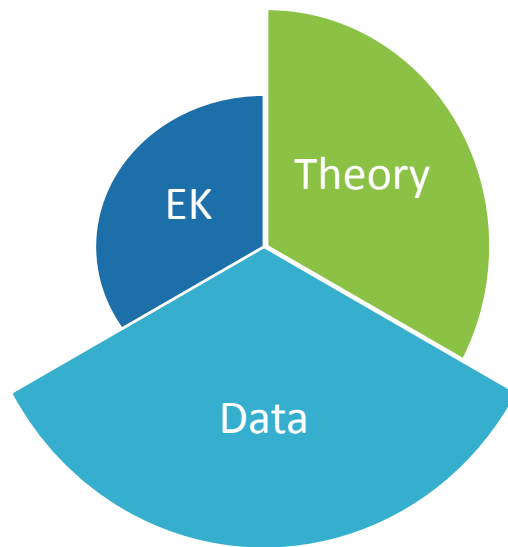
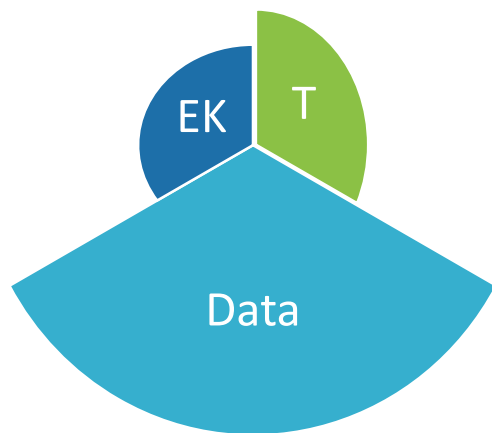
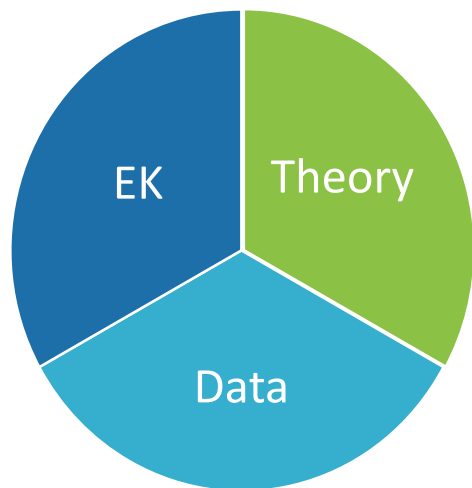
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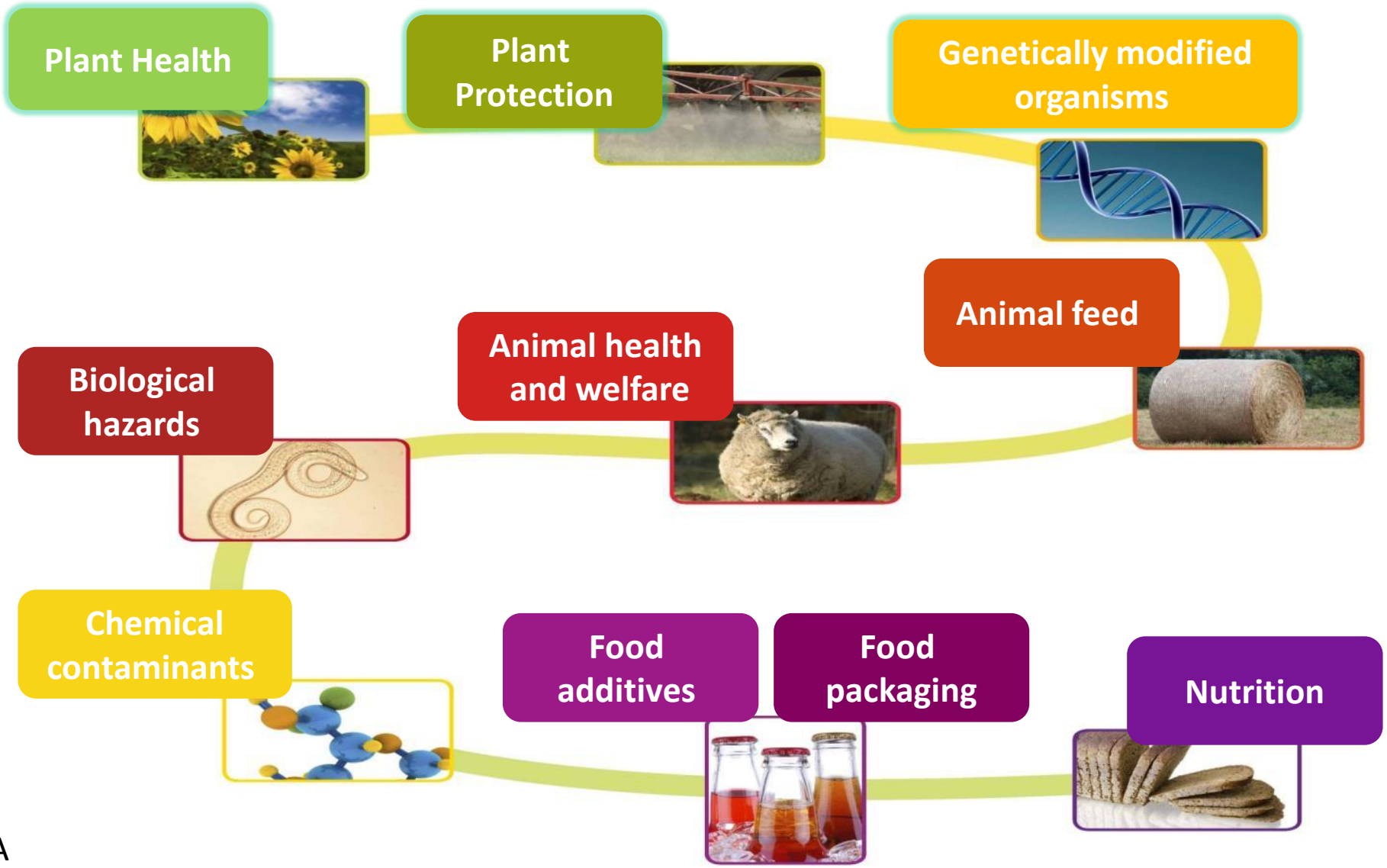
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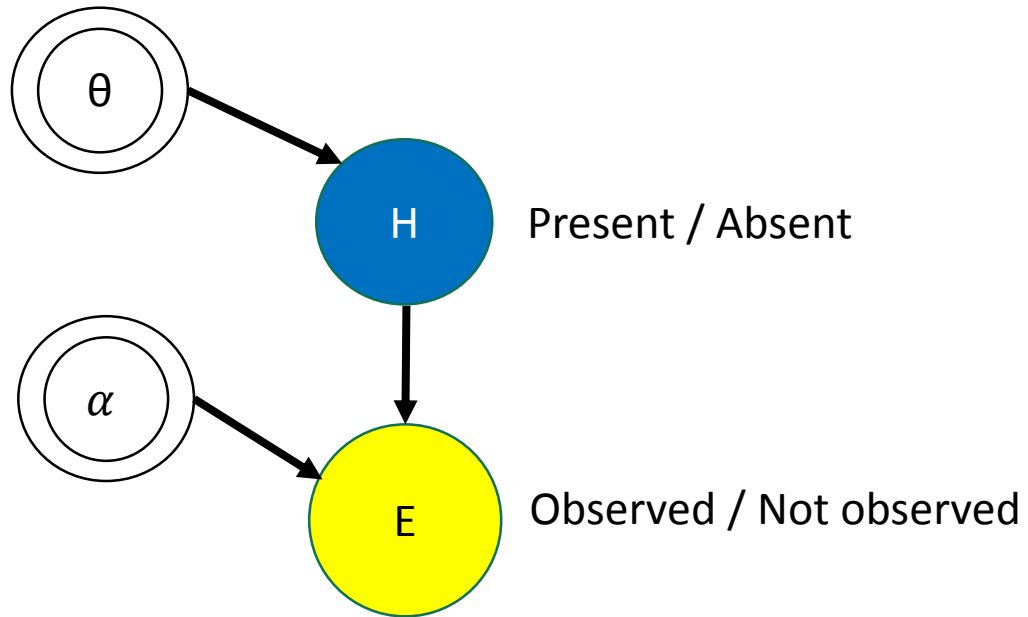




# European Food Safety Agency (EFSA)



# Is the crayfish still present?

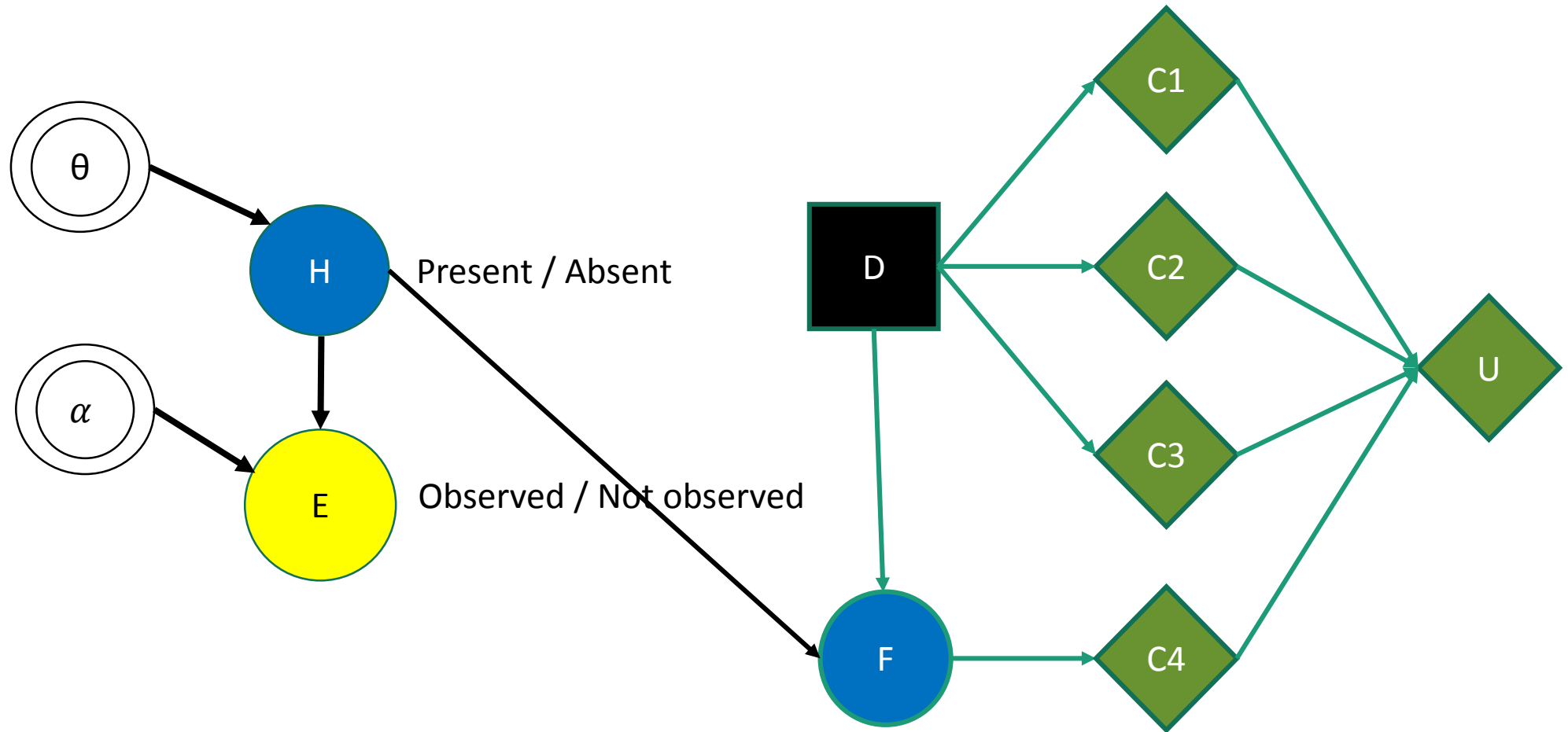


H	
Present	$\theta$
Absent	$1 - \theta$

	H	
E	Present	Absent
Observed	$\alpha$	0
Not observed	$1 - \alpha$	1

$$P(H|\text{not } E) = \frac{(1 - \alpha)\theta}{1 - \alpha\theta}$$

# What to do with the crayfish?

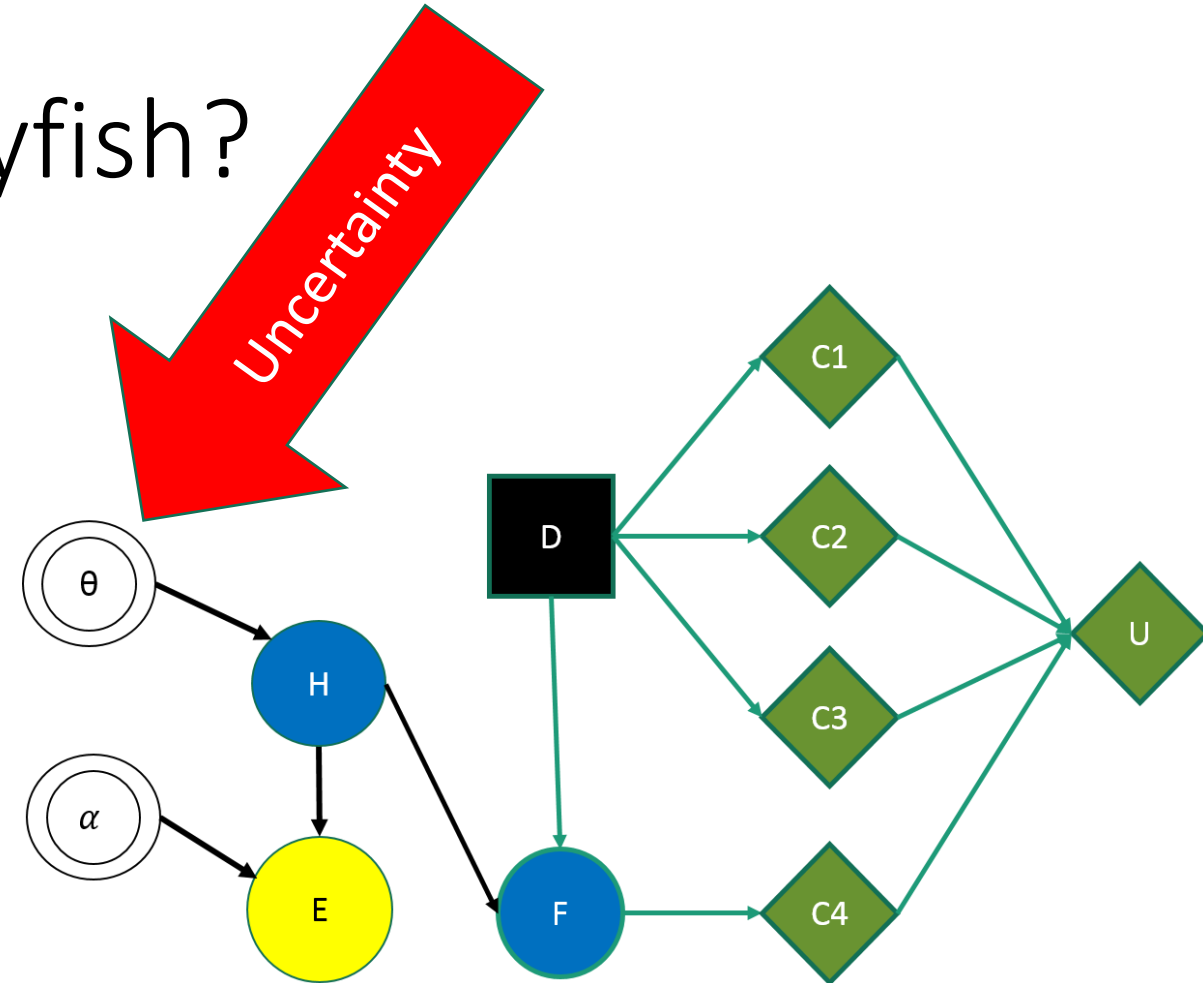


# What to do with the crayfish?

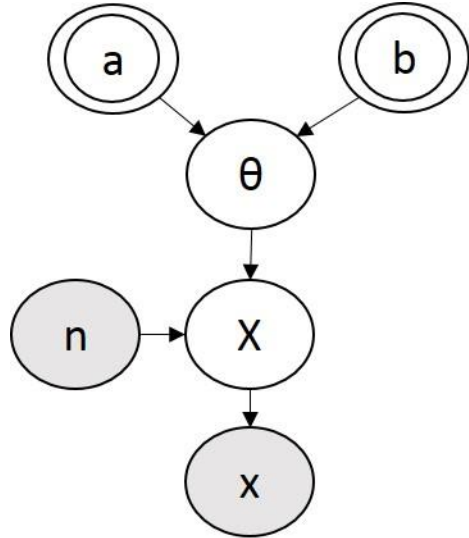


Management alternative	C1: Cost	C2: Neg Impact	C3: Acceptance
Do nothing	0	0	0
Mechanical removal	10	2	10
Add poison	5	10	2

C4: Loss is even worse if crayfish is present after management



# A simple Bayesian analysis



$\theta$  parameter of interest

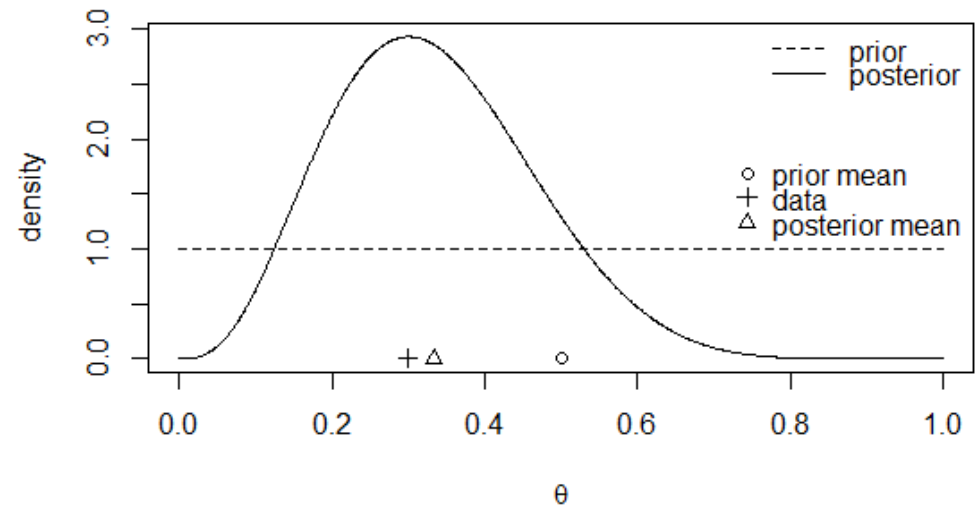
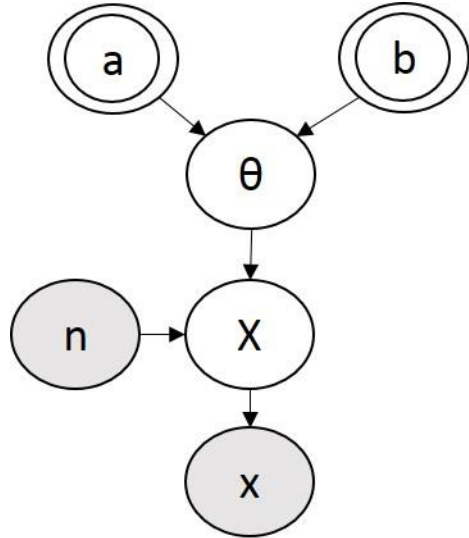
a & b expert knowledge on  $\theta$

X system variable

n sample size

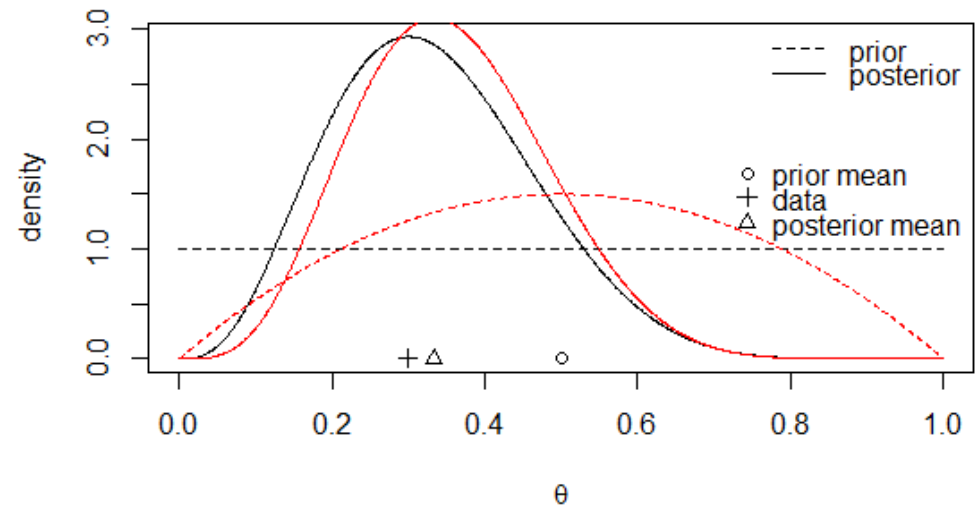
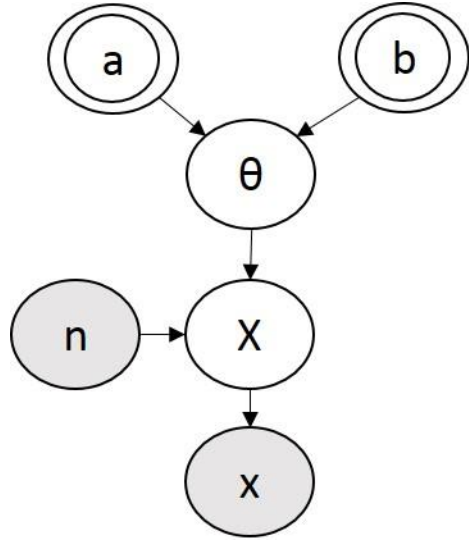
x observations of the variable

# A simple Bayesian analysis

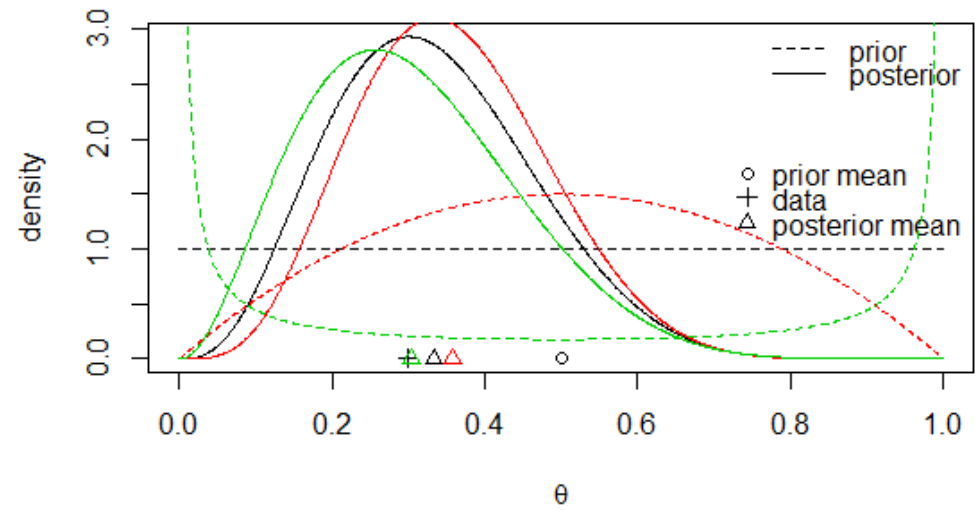
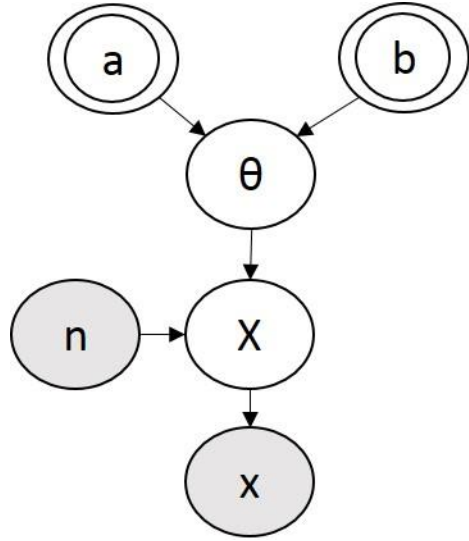




# A simple Bayesian analysis

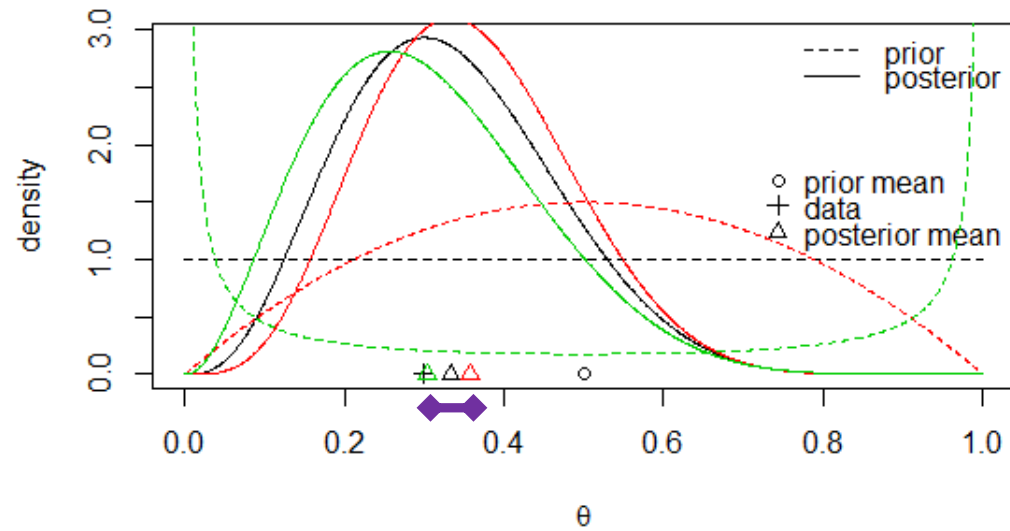


# A simple Bayesian analysis

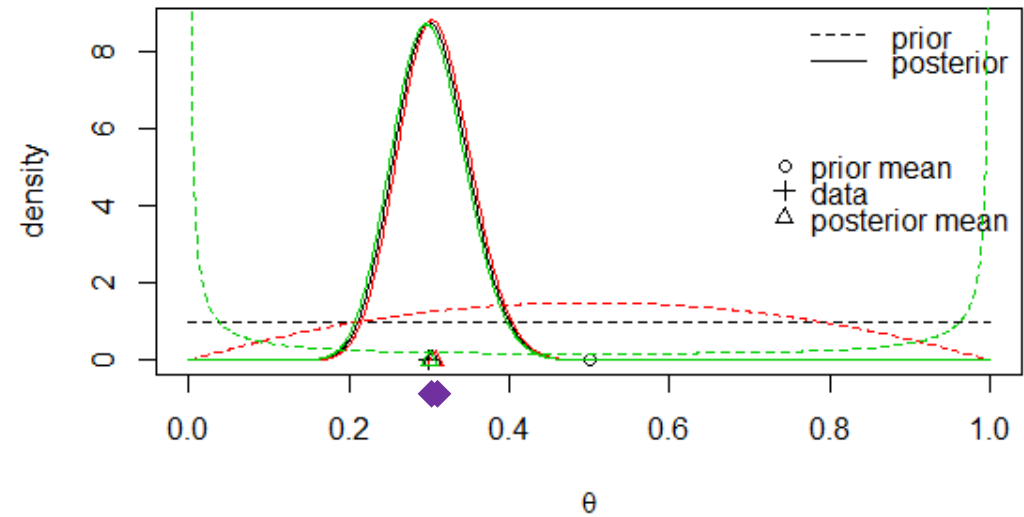


# The prior may matter – better get it right

## Small sample size



## Large sample size





# EK in risk and decision analysis

Quantitative risk models should be informed by systematically reviewed scientific evidence, however, in practice empirical evidence is often limited: in such cases it is necessary to turn to expert judgement.

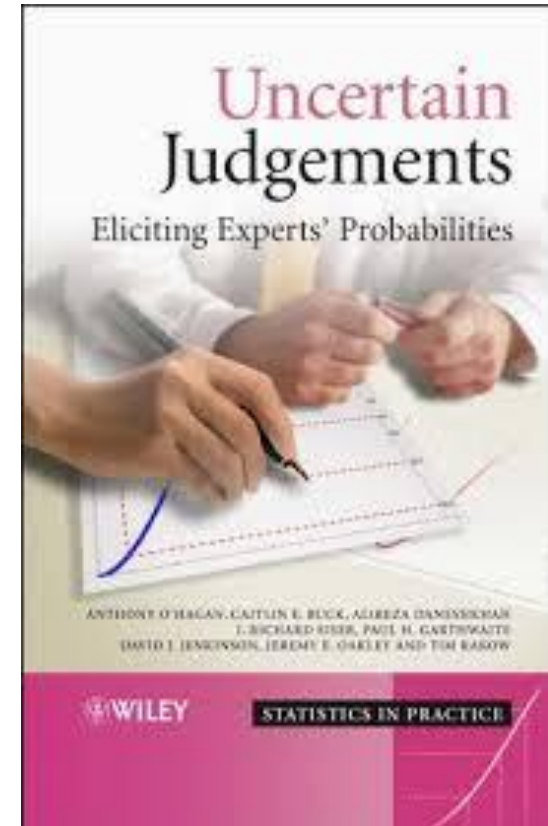
Psychological research has shown that unaided expert judgement of the quantities required for risk modelling - and particularly the uncertainty associated with such judgements - is often biased, thus limiting its value.

Accordingly methods have been developed for eliciting knowledge from experts in as unbiased a manner as possible.

<https://www.efsa.europa.eu/en/press/news/140623>

# Expert's Knowledge Elicitation

- Aim to describe the Expert's Knowledge about one or more uncertain quantities in probabilistic form
- i.e. a joint probability distribution for the random variable in question
- EKE can be used to build priors distributions or prior predictive distributions





# An Expert Knowledge Elicitation

- Formulate the elicitation questions
- Ask experts about
  - Probabilities
  - Quantiles
  - Probability intervals
  - Moments or other descriptions of a probability distribution
- Fit and aggregate into a probability distribution for the uncertain quantity

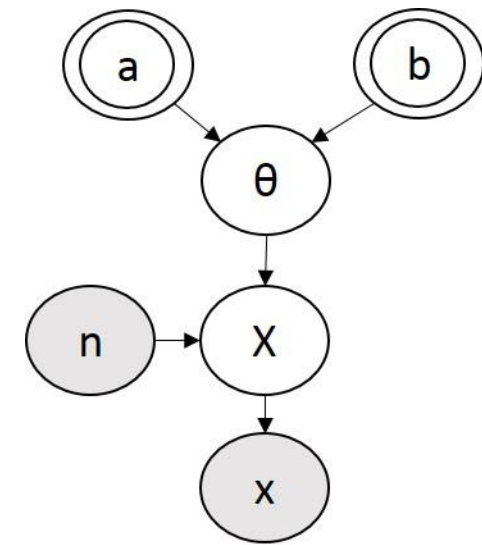


# Direct methods for EKE

- Simple and a bit crude
  - *Intervals* – Lower and Upper limits, then a Uniform distribution
  - *Triangular distributions* – Mode, Lower and Upper limits
- Cumulative Density Function (CDF)
  - *Quartiles* – 4 intervals, median and 25th and 75th percentiles
  - *Tertiles* – 3 intervals with equal probability
  - *Probabilities/Hybrid* – Choose probabilities and intervals
- Probability Density Function (PDF)
  - Mode/Mean, percentiles, shape,...
  - Place chips, draw it by hand...



# Indirect methods for EKE



- Equivalent Prior Sample (EPS)

- *What is the expected frequency of the event?*
- *What is the size a sample that you imagine to have behind this estimate?*

$$\frac{x}{n} = ? \quad n = ?$$

- Hypothetical Future Sample (HFS)

- *In a future sample of size 100 – in how many times has the event occurred?*

$$n = 100 \quad x = ?$$





# Selection of Structured EKE Software

- EXCALIBUR (EXpert CALIBration): [www.lighttwist.net/wp/excalibur](http://www.lighttwist.net/wp/excalibur)
- ElicitN: [www.downloadcollection.com/elicitn.htm](http://www.downloadcollection.com/elicitn.htm)
- SHELF (The SHEffield ELicitation Framework): [www.tonyohagan.co.uk/shelf/](http://www.tonyohagan.co.uk/shelf/)
- MATCH Uncertainty Elicitation  
Tool: [optics.eee.nottingham.ac.uk/match/uncertainty.php#](http://optics.eee.nottingham.ac.uk/match/uncertainty.php#)
- UncertWeb - The Elicitor: <http://elicator.uncertweb.org/>
- Variogram elicitation: [www.variogramelicitation.org](http://www.variogramelicitation.org)
- Unicorn: [www.lighttwist.net/wp/unicorn-download](http://www.lighttwist.net/wp/unicorn-download)

# An example

– elicit the probability of the crayfish individuals to survive the winter



- The SHELF R-package
- A web-interface for the SHELF R-package:  
[optics.eee.nottingham.ac.uk/match/uncertainty.php#](http://optics.eee.nottingham.ac.uk/match/uncertainty.php#)
- Roulette
- Quartiles
- Tertile

```
EK.R x
1 library(SHELF)
2
3 ## elicit one expert or consensus distribution
4 EK_info <- roulette(lower = 0, upper = 100, gridheight = 10, nbins = 10)
5
6 EK_info
7 $v
8 [1] 10 20 30 40 50 60 70 80 90 100
9
10 $p
11 [1] 0.1428571 0.5714286 0.8571429 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
12 [10] 1.0000000
13
14 ## fit distribution to expert info
15 EK <- fitdist(vals = EK_info$v, probs = EK_info$p, lower = 0, upper = 100)
16
17 plotfit(EK, ql = 0.05, qu = 0.95, d = "beta")
18
19 |
```

C:/Users/Ullrika/Box Sync/Bayesian/bayes@lund2017 - Shiny  
http://127.0.0.1:6563 Open in Browser Publish

# Roulette elicitation

Show fit  
 Spread end probs over empty bins

Distribution

- Normal
- Student t
- Gamma
- Log normal
- Log Student t
- Beta
- Best fitting

Student-t degrees of freedom  
3

lower feedback quantile  
0.05

upper feedback quantile  
0.95

Finish

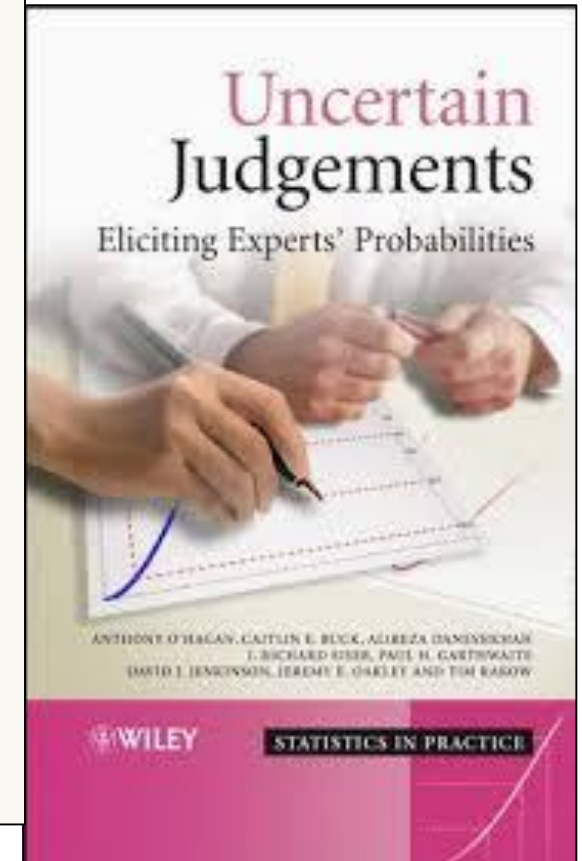
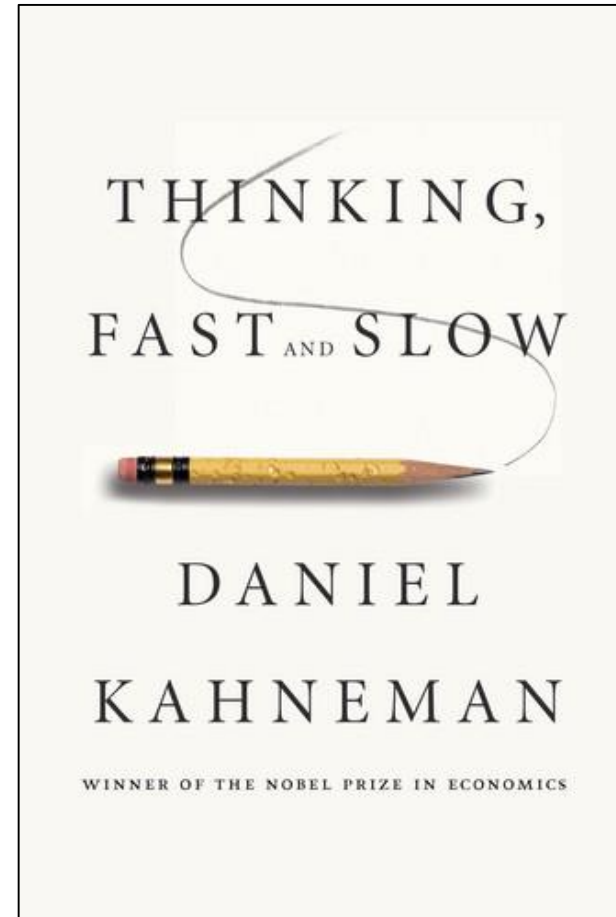
**Total probs: 7**

Bin Range	Count
0-10	1
10-20	3
20-30	2
30-40	1
40-50	1
50-60	0
60-70	0
70-80	0
80-90	0
90-100	0

**Beta(3.55, 14.6)**

# Psychological factors and elicitation

- Anchoring and adjustment
- Availability
- Range–frequency compromise
- Representativeness and baseline neglect
- Conjunction fallacy
- The law of small numbers
- Overconfidence



# Elicitation with multiple experts

- Psychological factors when working with several experts
- Behavioural aggregation
  - Group elicitation
  - One or several iterations, individually and in group
- Mathematical aggregation
  - Treat each expert's distribution as data and update the decision maker's belief
  - Pooled opinions – linear or logarithmic pooling
  - Calibrate experts and weight according to their performance





# Alternative protocols for EKE

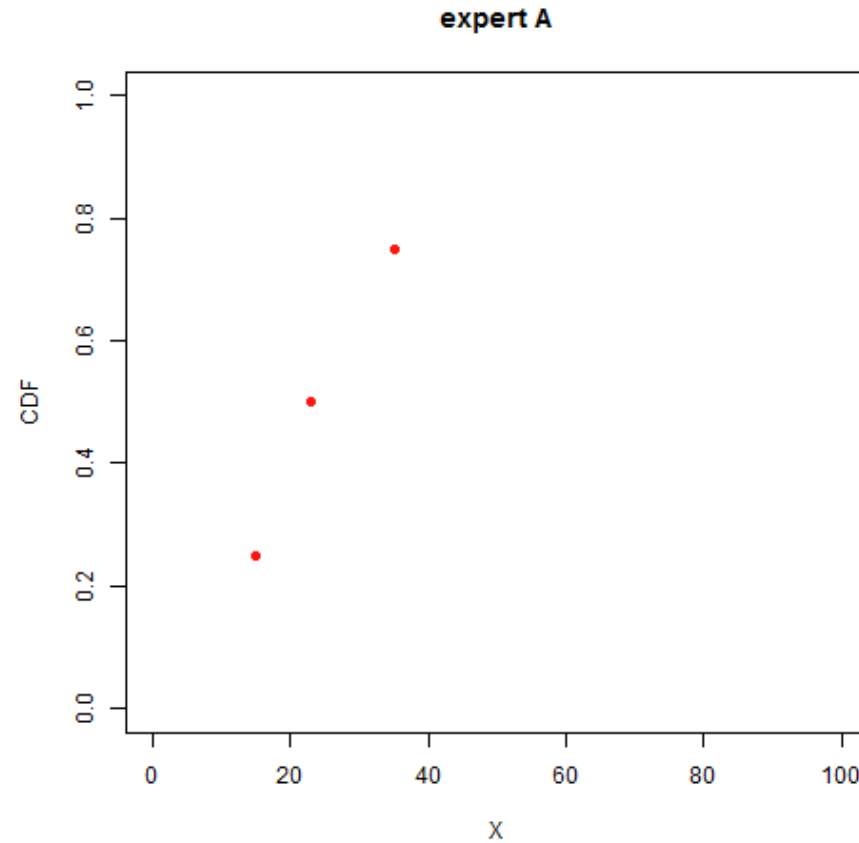
- the Sheffield protocol with group interaction of experts, consensus distributions
- the Cooke protocol with use of seed questions for the calibration of experts, no interaction
- a Delphi protocol on written expert elicitation with feedback loops, anonymous sharing of the results between iterations



An example  
– elicit the probability of the crayfish individuals to survive the winter



75% percentile  
Median  
25% percentile





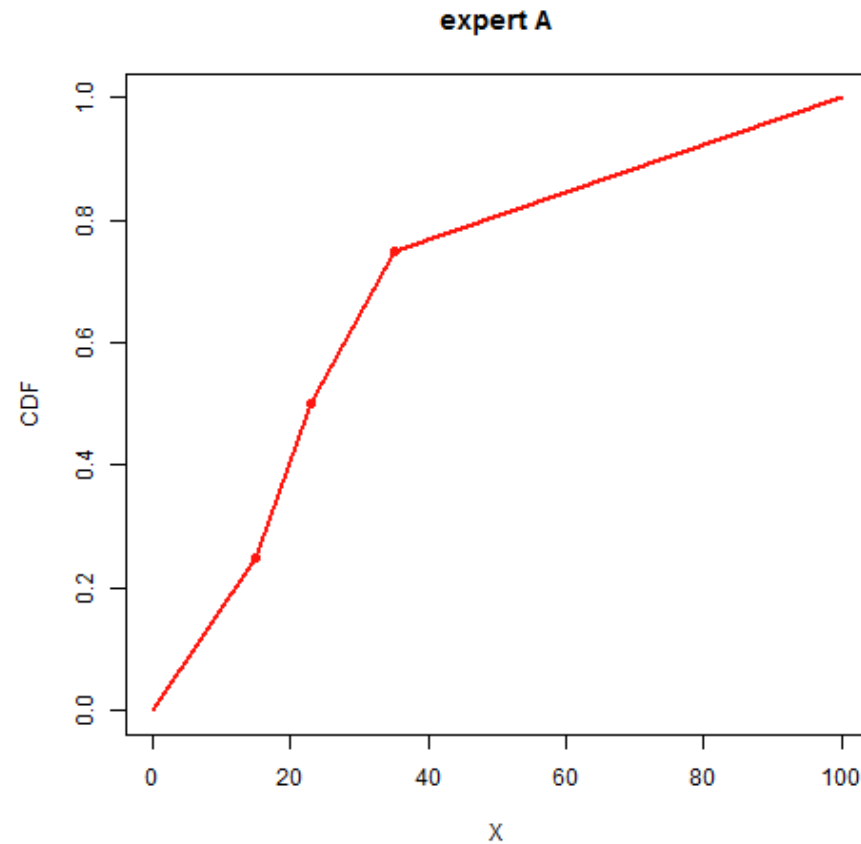
An example  
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75% percentile

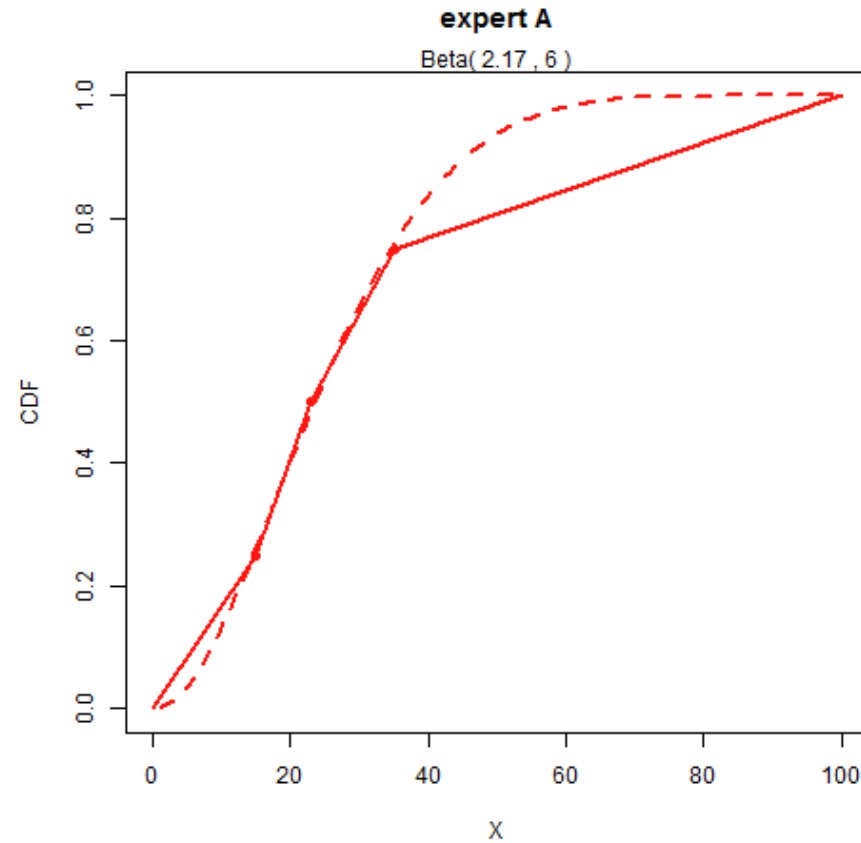
Median

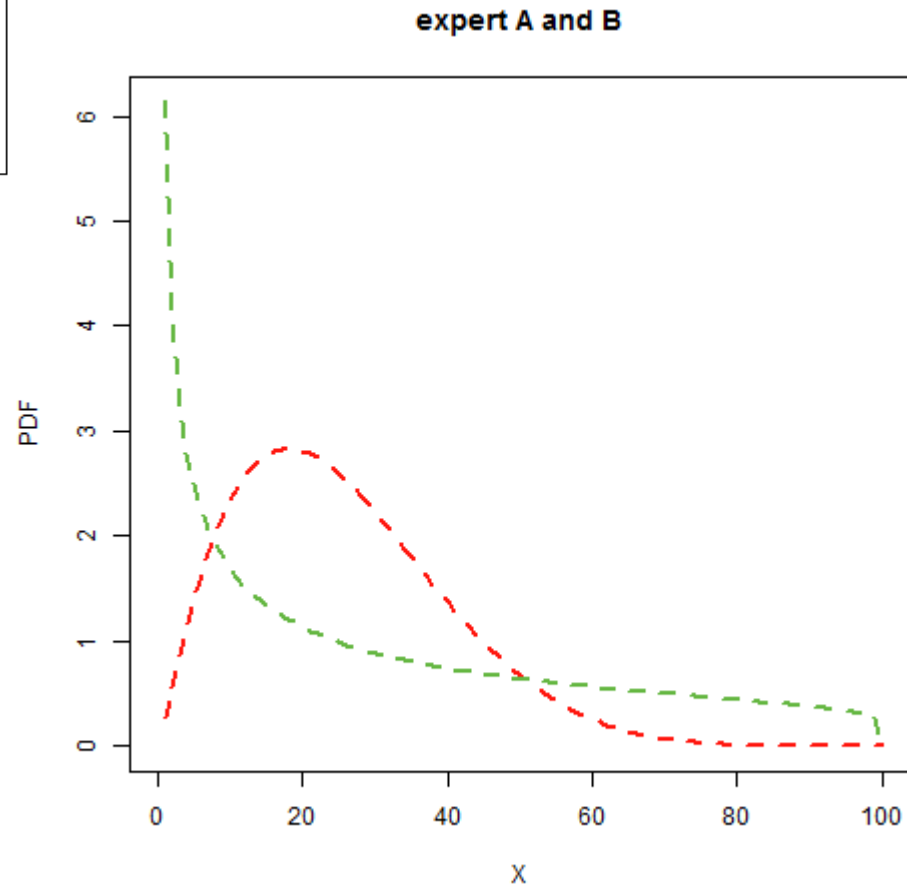
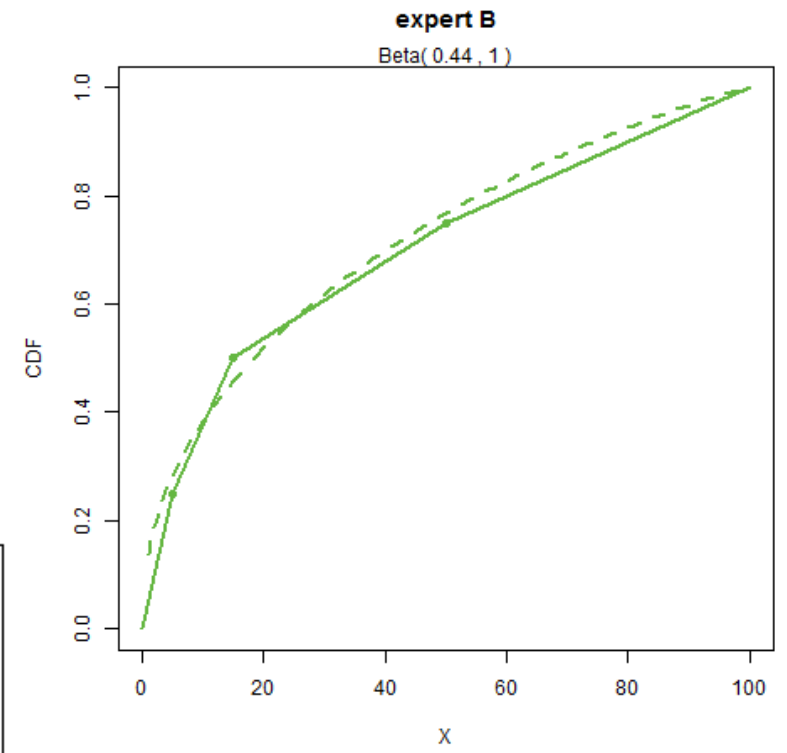
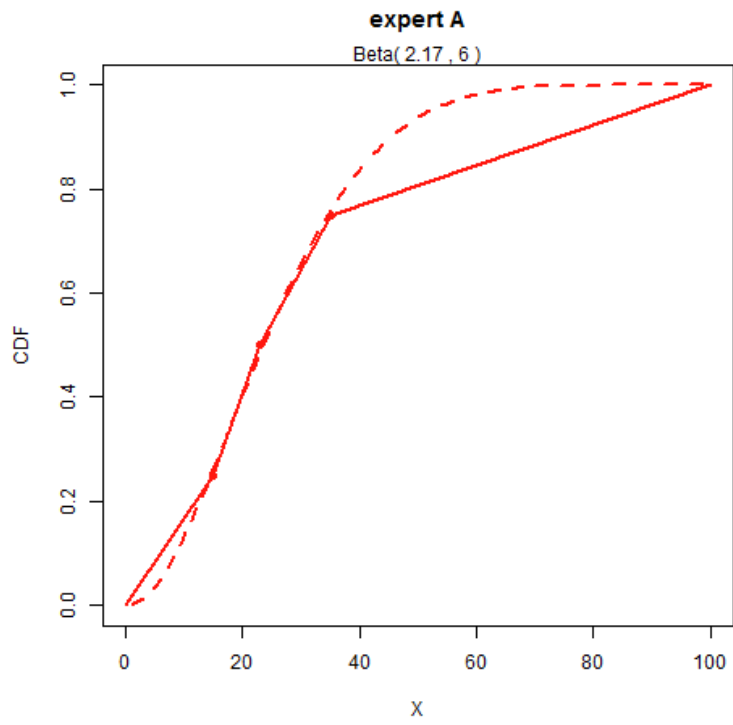
25% percentile

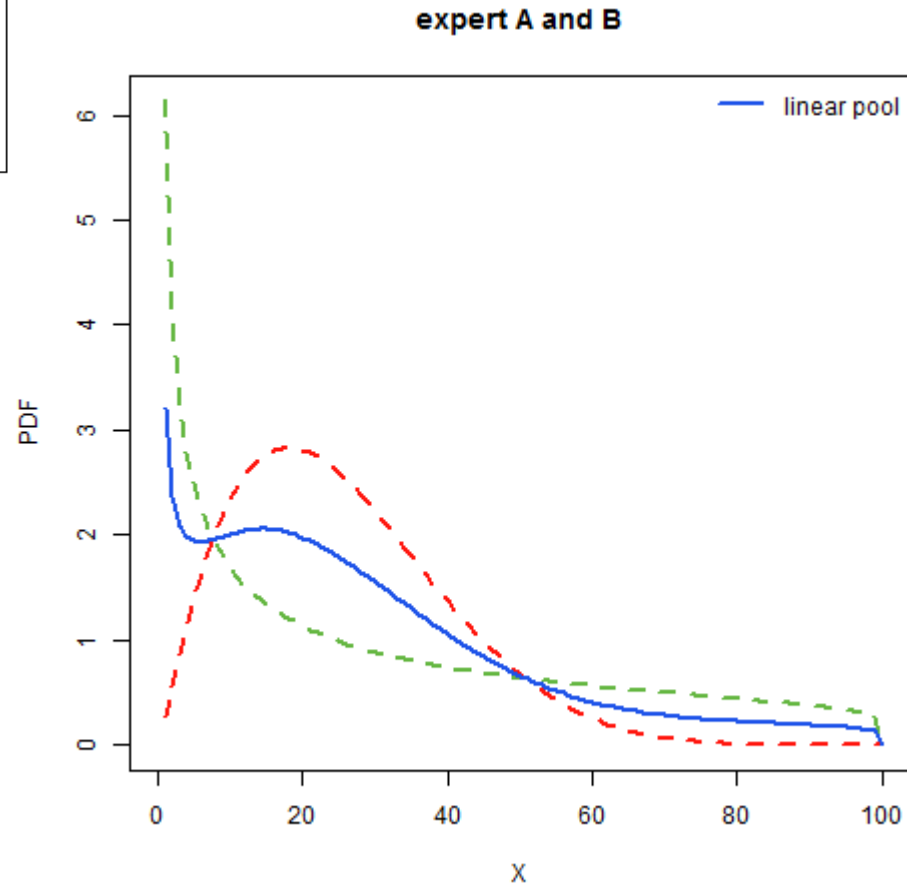
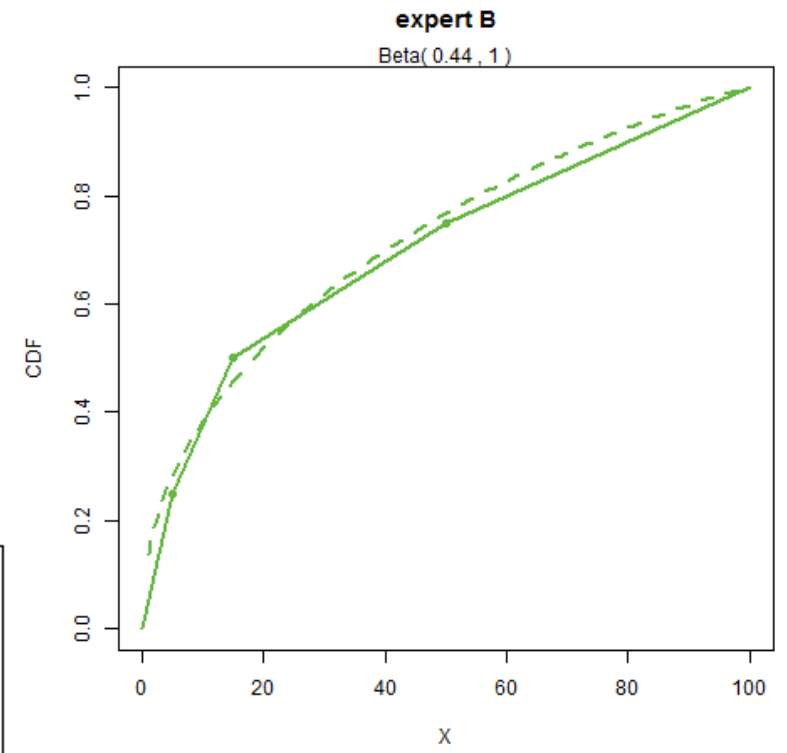
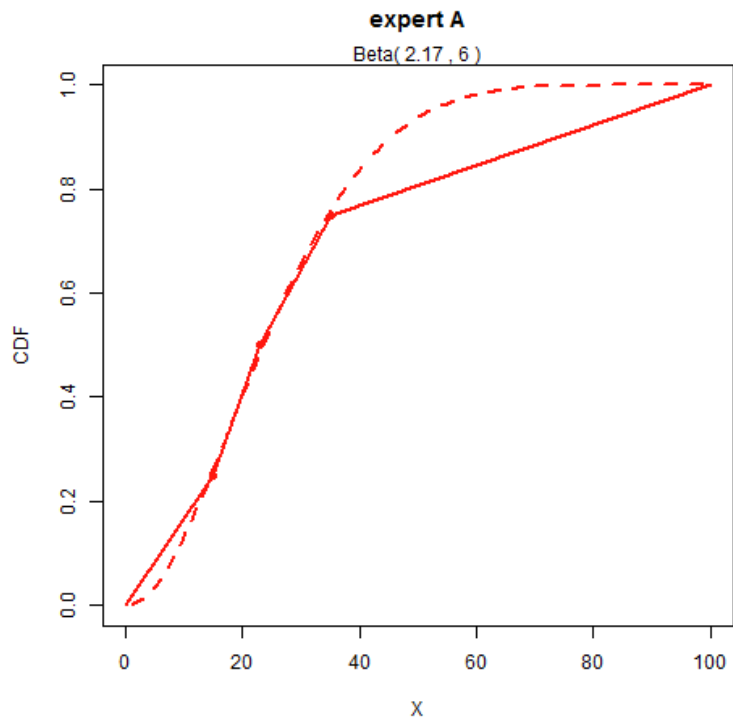




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# Train the experts in making probabilistic judgments

- Get them custom to use probability density functions as a device for representing subjective uncertainty
- Clarify what is needed for the assessment, what are the uncertain quantities and how they are used to assess risk
- Reassure the experts understand that they will not be expected to claim certainty they do not have
- ★ • Encourage experts to be honest
- Give the experts a practice elicitation exercise
- Discuss psychological biases







# The benefit of quantifying uncertainty using probability

- X is the proportion surviving the first spray
- Y is the proportion surviving the second spray
- Proportion surviving both applications is  $XY$

X is small and Y is small, what is then  $XY$ ?

What does "small" mean?





# The benefit of quantifying uncertainty using probability

- X is the proportion surviving the first spray
- Y is the proportion surviving the second spray
- Proportion surviving both applications is  $XY$

X and Y without uncertainty result in a single value of  $XY$

A false sense of security







# The benefit of quantifying uncertainty using probability

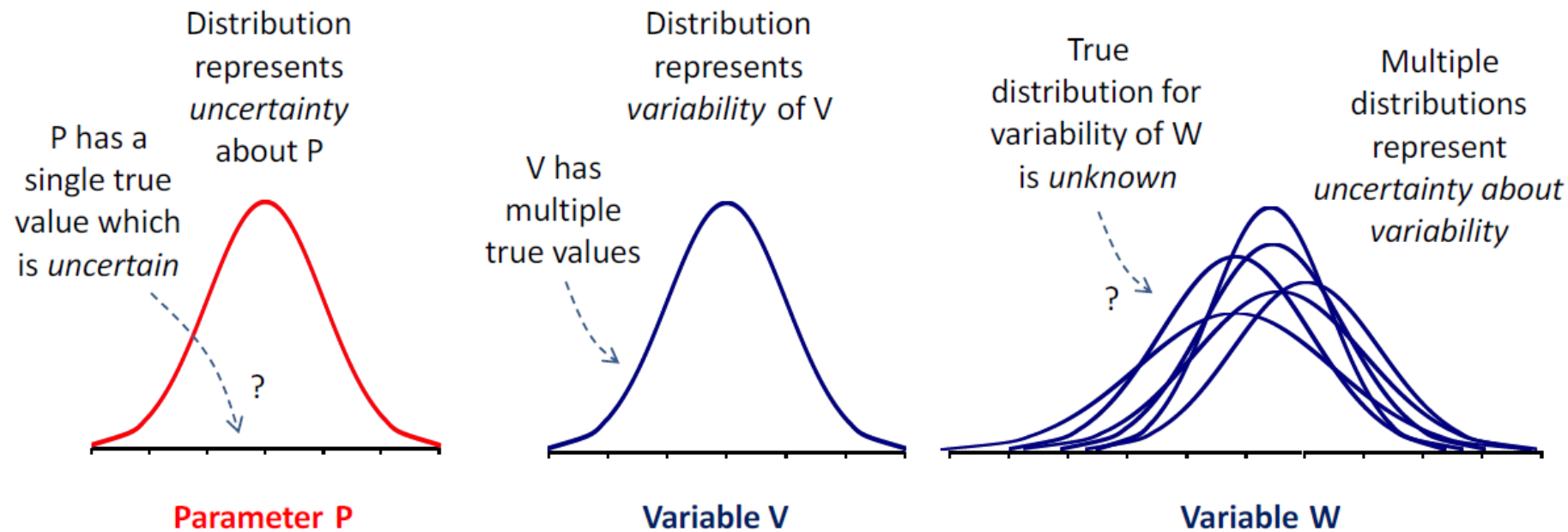
- X is the proportion surviving the first spray
- Y is the proportion surviving the second spray
- Proportion surviving both applications is XY

Using point estimates in input can result in biased estimate of overall risk

Plug in estimates – no uncertainty	Consider uncertainty in inputs
$E(X) = 2\%$	$X \sim U(0,4\%)$
$E(Y) = 2\%$	$Y \sim U(0,4\%)$
$XY = E(X)E(Y) = 0.04\%$	$E(XY) = 0.053$



# Aleatory and epistemic uncertainty



**Figure 2:** Illustration of the distinction between uncertainty and variability (left and central graphs), and that both can affect the same quantity (right hand graph).